#### UNITED STATES DISTRICT COURT SOUTHERN DISTRICT OF NEW YORK

IN RE:

25-md-3143 (SHS) (OTW)

OPENAI, INC., COPYRIGHT INFRINGEMENT LITIGATION

Hon. Sidney H. Stein Hon. Ona T. Wang

This Document Relates To:

All Actions

NOTICE OF OPENAI'S TECHNOLOGY TUTORIAL DEMONSTRATIVES



# OpenAl's Technology Tutorial

In re OpenAl Copyright MDL MDL No. 25-md-3143 (S.D.N.Y.)

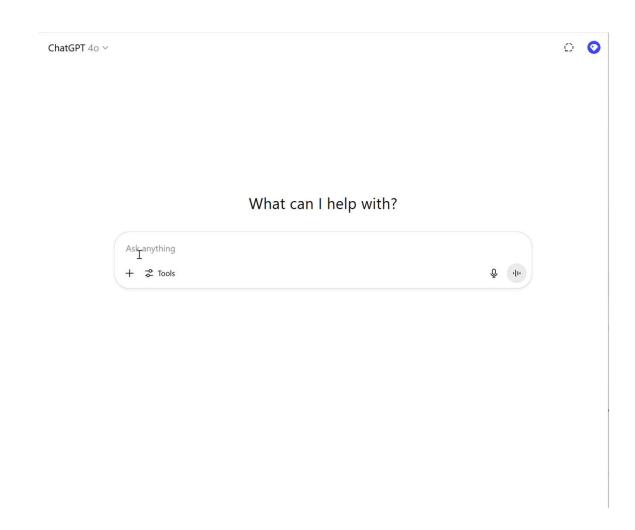
June 26, 2025

#### What is a GPT?



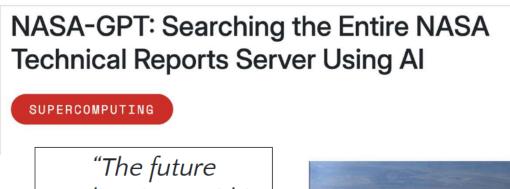






#### **ChatGPT generates innovation**





"The future applications within NASA are nearly endless."



### ChatGPT uses Large Language Models (LLMs): "GPTs"

A GPT is a statistical model that predicts the next word by learning patterns in language

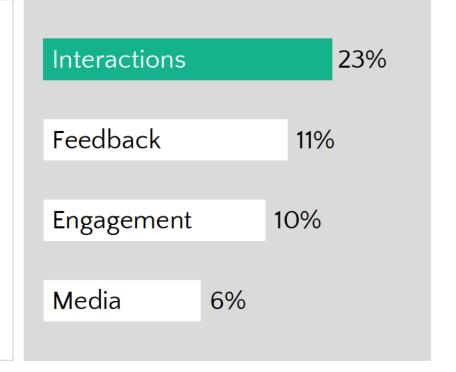
#### **Prompt:**

How do humans learn language?

#### Response:

Through patterns, observation, and social

\_\_\_\_



#### What is a GPT?



Not a database; generates new content

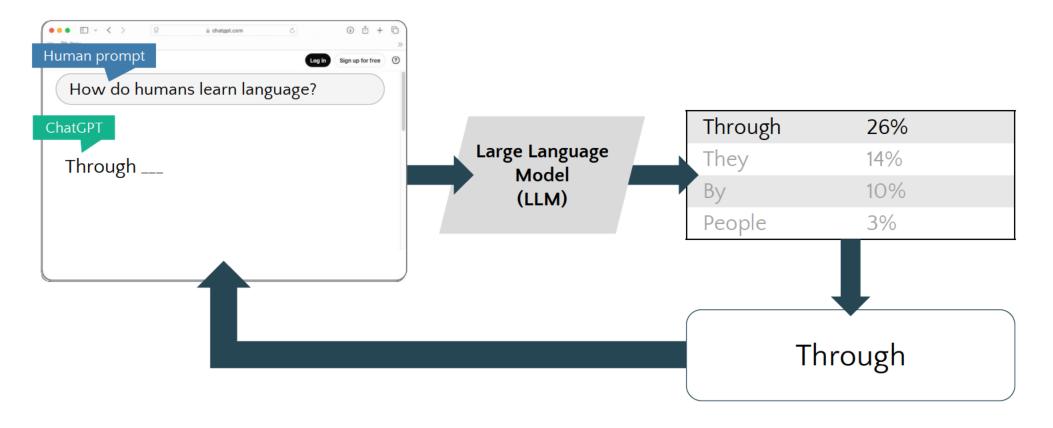
P RE-TRAINED

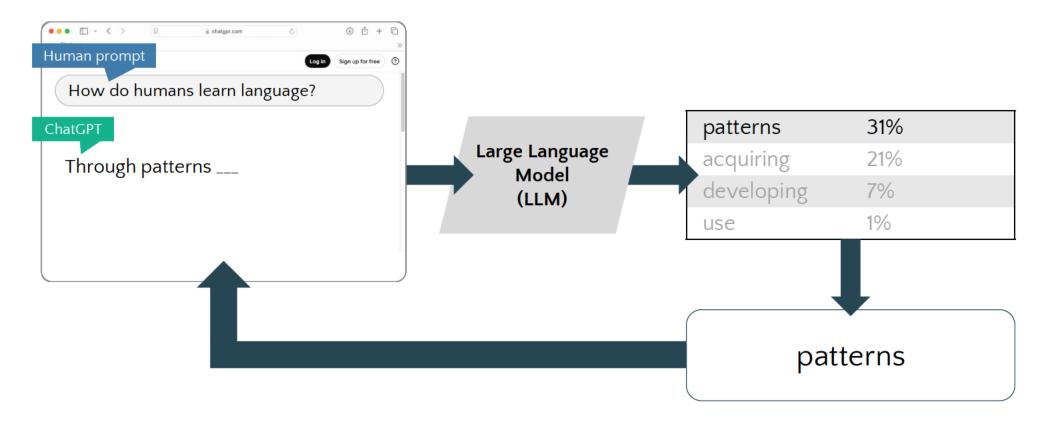
- \_\_ **Trained** to identify linguistic **patterns**
- T RANSFORMER
- \_ **Transforms** language data into a platform for **innovation**

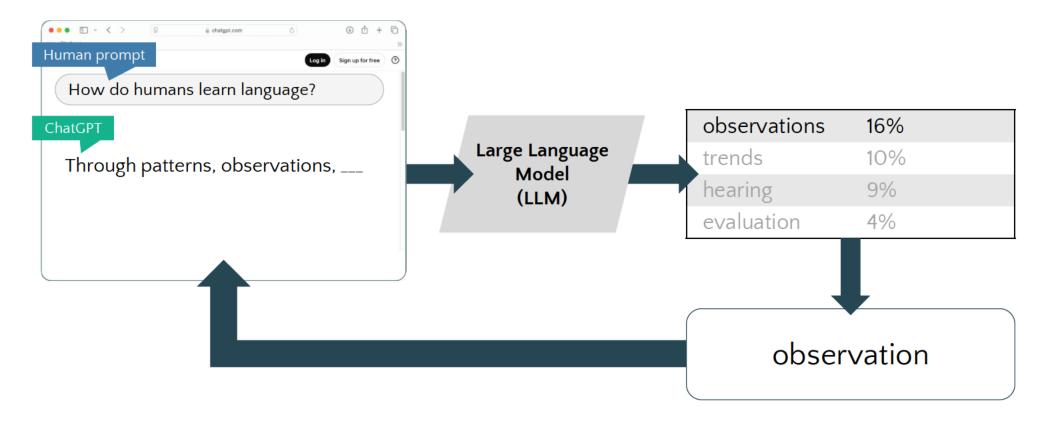
#### LLMs are not databases or libraries

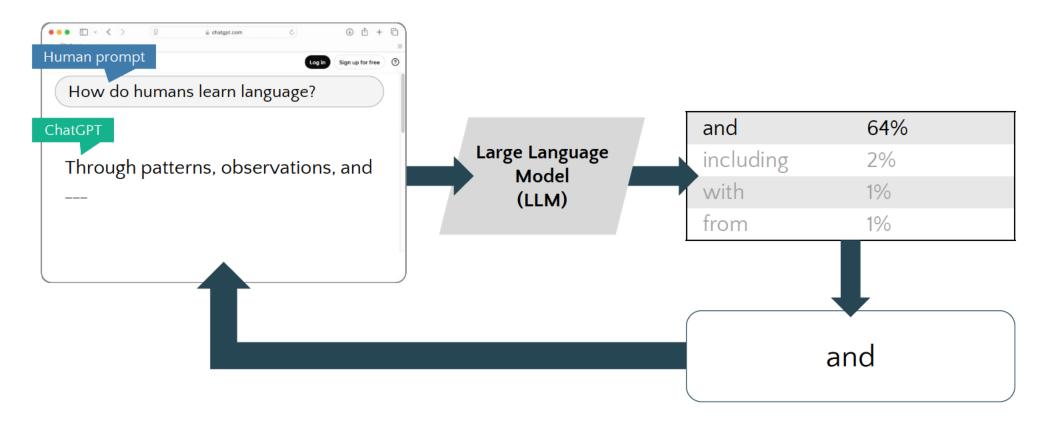
They do not store information or retrieve information from a database

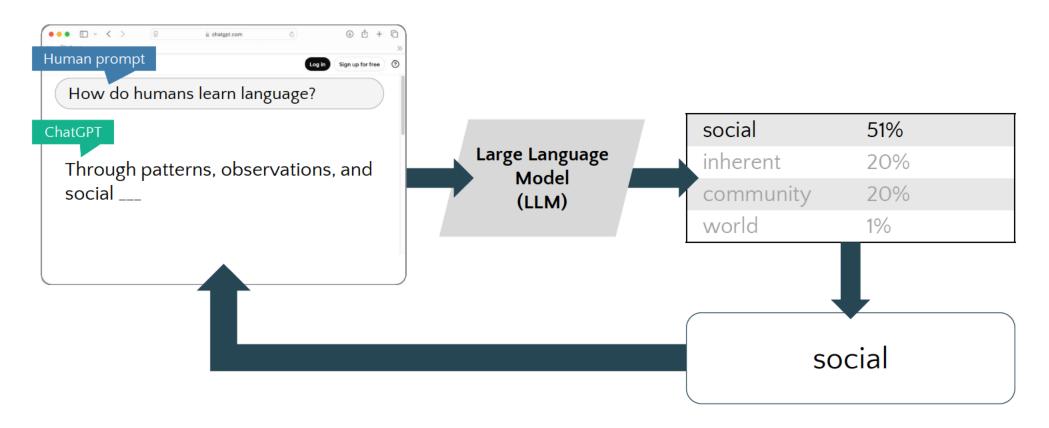


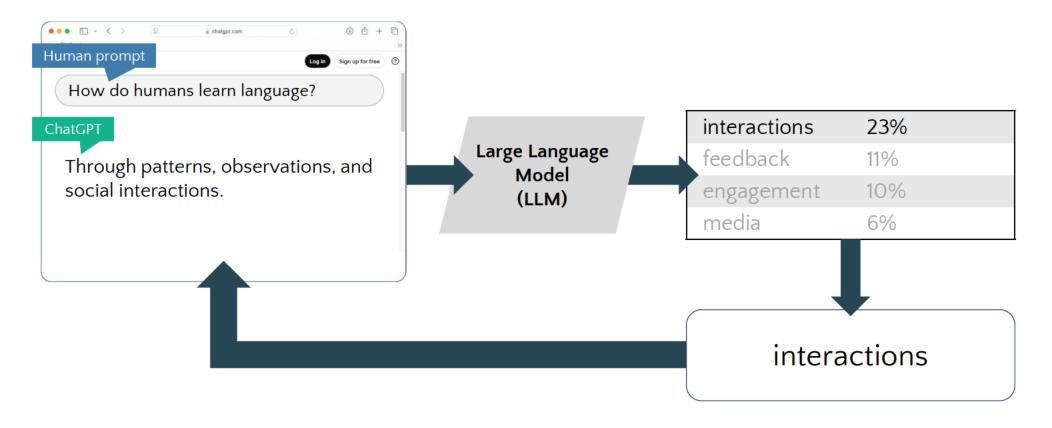




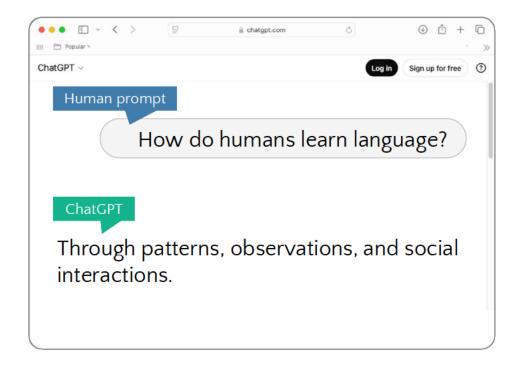


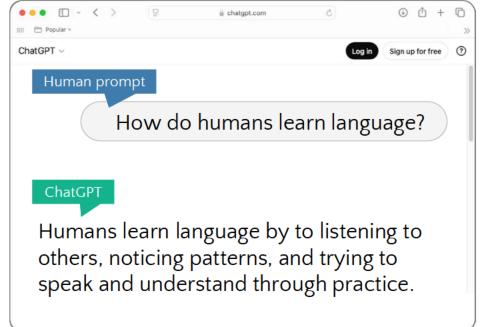






# Because LLMs are not designed to memorize, the same input can result in different outputs





#### What is a GPT?

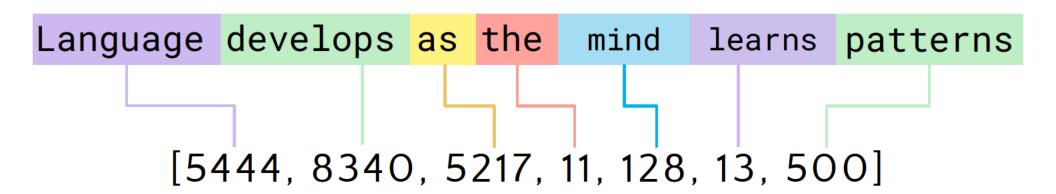


\_\_ Not a database; generates **new** content

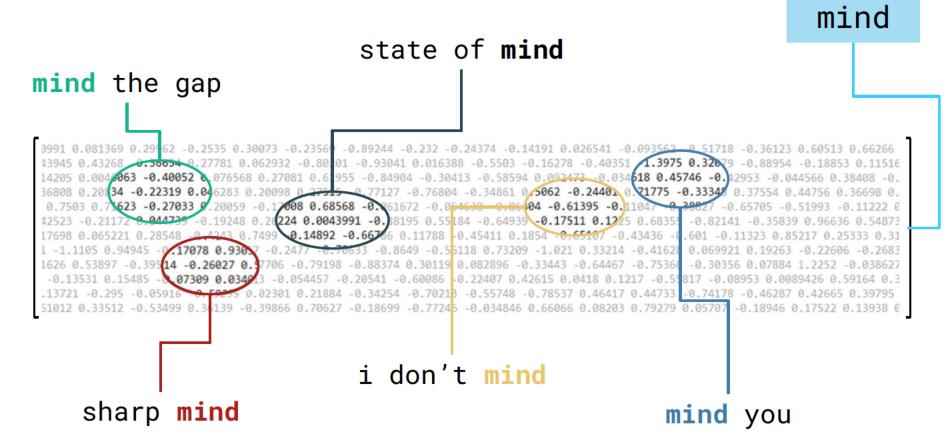
P RE-TRAINED

- Trained to identify linguistic patterns
- T RANSFORMER
- \_ **Transforms** language data into a platform for **innovation**

# During tokenization, words are transformed into numbers that a computer can understand



#### Each word = thousands of numbers



#### Each word = thousands of numbers

brain

```
-0.65099 1.26331 -0.87651 -1.72957 1.97517 -0.15320 0.11393 -0.84078 0.63132 1.76751 -1.89129 0.74010 0.11836 -0.67188 -1.77762
7 0.58466 0.42617 0.31083 -1.45158 -0.61226 -1.20194 0.27056 1.19842 -1.71352 -1.15130 0.78276 -1.98115 0.24613 0.77700 -0.358495 1.43668 -0.64877 0.40269 -1.26063 -1.25985 -0.23417 -0.58839 0.57958 0.85748 0.20365 -1.83983 -1.79238 -0.01656 1.11073 -1.8417. 1.44323 1.31257 0.54199 0.29543 -1.70525 1.79376 0.91679 -0.15764 1.35187 -0.10371 1.57899 0.28532 0.84626 -1.44230 0.02376 -1. -0.83522 -1.83013 0.07620 1.02852 0.59177 -1.29791 -1.89422 1.87762 -0.52827 -1.05487 1.98063 -1.53682 -1.27855 -0.80357 0.03492 0.59323 0.97371 -0.15144 -0.87792 0.01189 -0.47492 1.62576 0.89438 -0.62645 1.87121 -0.14262 -1.62334 -0.50791 -1.62259 -1.67865 0.55613 1.06068 -0.16104 0.59806 0.11356 -0.02945 1.05590 -1.27190 1.86503 -1.34049 0.01791 -1.63649 1.75626 1.60784 -1.68794 -1.53942 1.10862 0.17078 0.938227 -1.81691 0.33745 -0.91474 -0.25783 -0.72632 -1.39547 -1.65267 0.69399 0.57808 -1.15688 1.01593 -1.31560 1.77706 -0.39353 -1.31560 1.77706 -0.39363 -1.34649 0.039702 -1.66281 1.75238 -0.12592 0.65880 -0.93610 -0.49126 -1.35556 -0.68012 -0.93368 -1.8673 -1.31560 1.77706 -0.26027 0.9 9 -0.17338 0.42629 -0.02297 0.73985 -0.91590 0.59878 1.63638 -1.72895 -0.25977 -1.25119 -1.63073 -0.28878 -0.61623 -1.20134 0.27056 1.19842 -1.71352 -1.15130 0.78276 -1.98115 0.24613 0.77700 -0.35849 -1.00467 1.43668 -0.64872 -1.45158 -0.61226 -1.20134 0.27056 1.19842 -1.71352 -1.15130 0.78276 -1.98115 0.24613 0.77700 -0.35849 -1.00467 1.43668 -0.64872 -1.45158 -0.61226 -1.20134 0.27056 1.19842 -1.71352 -1.15130 0.78276 -1.98115 0.24613 0.77700 -0.35849 -1.00467 1.43668 -0.64872 -1.45158 -0.61226 -1.20134 0.27056 1.19842 -1.71352 -1.15130 0.78276 -1.98115 0.24613 0.77700 -0.35849 -1.00467 1.43668 -0.64872 -1.45158 -0.61226 -1.20134 0.27056 1.19842 -1.71352 -1.15130 0.78276 -1.98115 0.24613 0.77700 -0.35849 -1.00467 1.43668 -0.64872 -1.45158 -0.61226 -1.20134 0.27056 1.19842 -1.71352 -1.15130 0.78276 -1.98115 0.24613 0.7
```

quick **brain** 

#### The model learns statistical relationships between tokens

· The model is shown billions of sequences of tokens and asked:

"Given this sequence, what token is most likely to come next?"

· The model runs these tests over and over, billions of times.

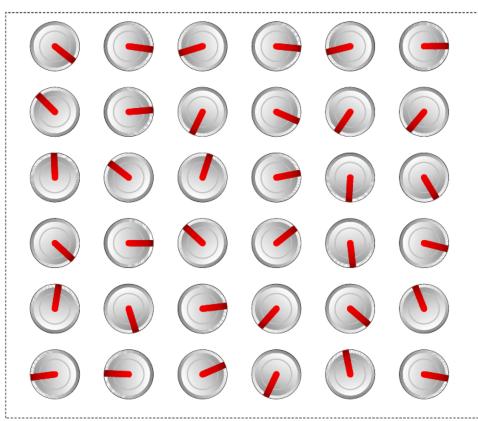
# Training is about adjusting the weights, not about memorizing training data

"...patterns emerge from repeated phrases..."

"...pathways strengthen with every utterance..."

"...babies observe linguistic patterns..."

"...babbling refines phoneme perception..."



Weights

**Training Data** 

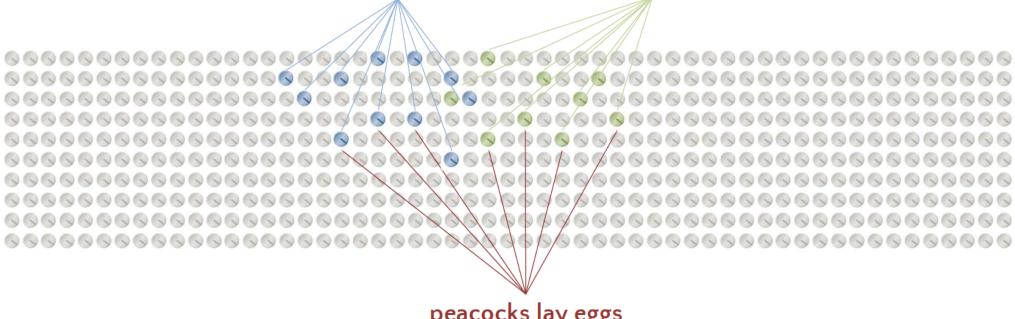
20

### Goal: teach the model to infer concepts not seen before

- "...most birds lay their eggs in carefully constructed nests..." "...a birds will lay three to five eggs per clutch..." "...unlike mammals, birds reproduce by laying eggs..."
- "...peacocks are birds in the pheasant family..."
- "...<mark>peacocks</mark> can fly short distances like other large birds..."
- "...peacocks are the most beautiful birds..."

birds lay eggs

peacocks are birds



#### The model creates something new

At Yale where ivy shadows fall,

No letters carve a judgment wall.

No A or B to stake your claim,

But Honors, Pass, a gentler name.

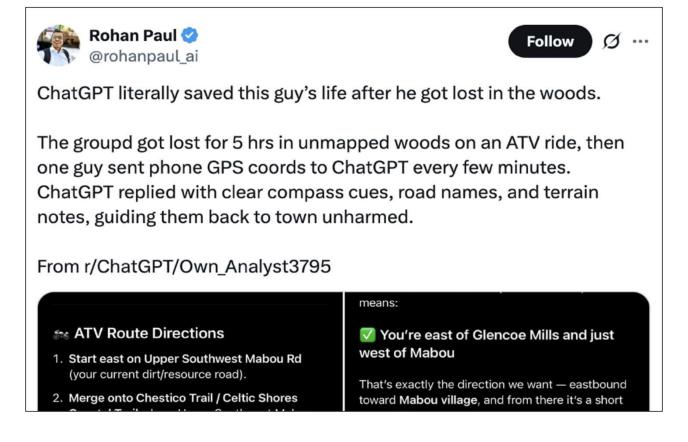
In halls where Socrates might teach,

The law is more than marks to reach.

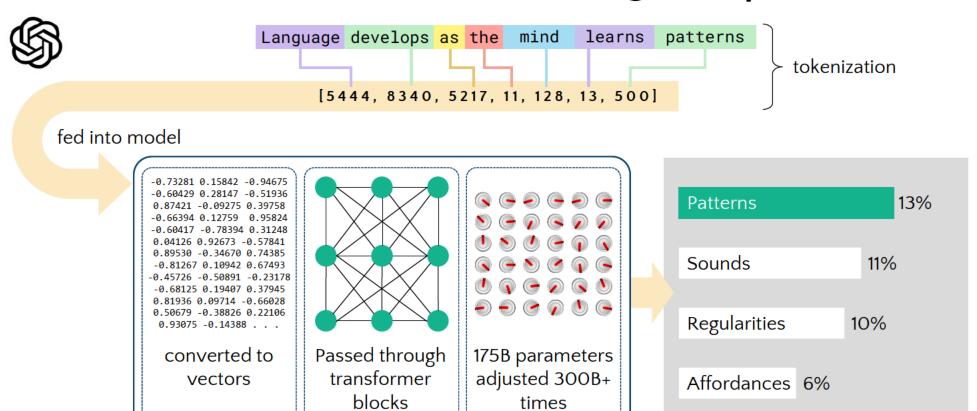
No race for rank, no bitter score,

Just minds that stretch, and learn, and soar.

#### ChatGPT solves real-world problems



### The model transforms text data and gleans patterns



#### What is a GPT?



\_\_ Not a database; generates **new** content

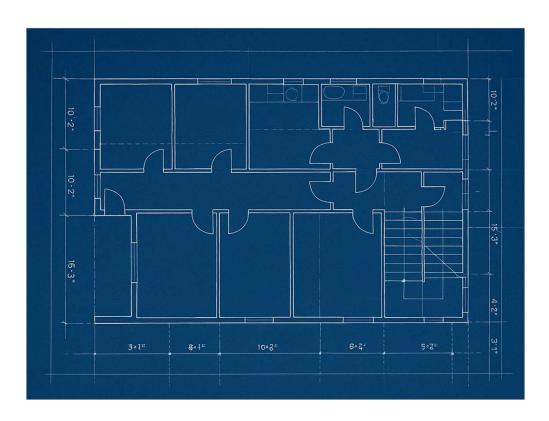


\_\_ **Trained** on a **massive** and **diverse** corpus of language data

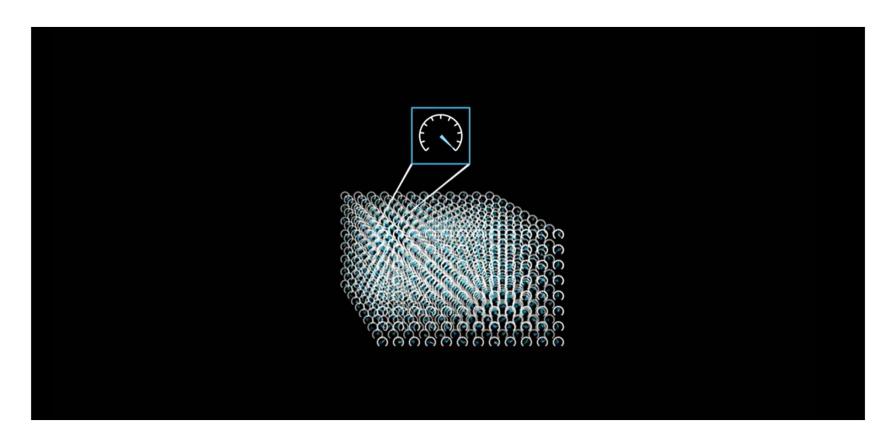


\_\_ **Transforms** vast amounts of data into a platform for **innovation** 

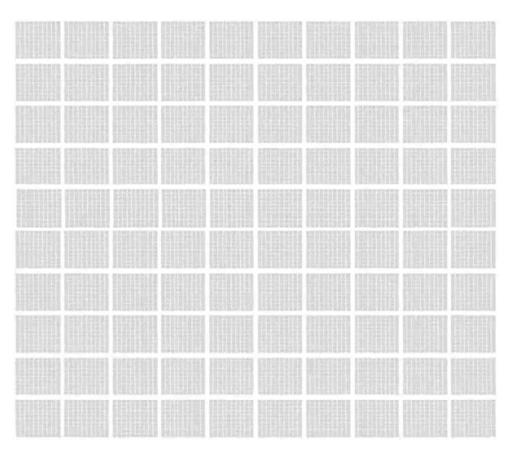
### Model Architecture



# Billions (or trillions) of parameters (weights) are organized into a model architecture.

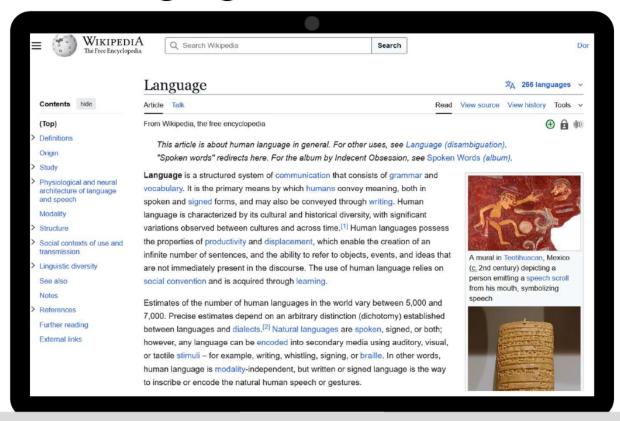


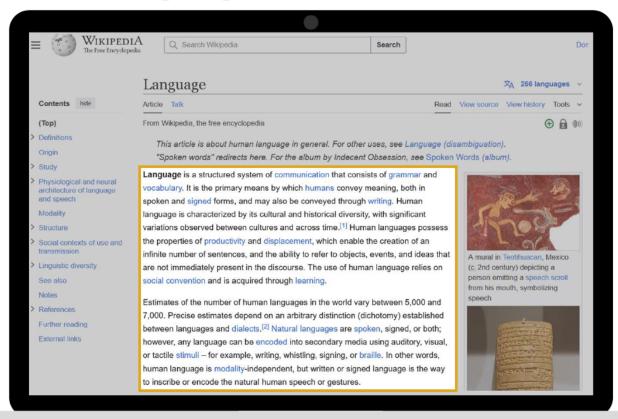
# Training an LLM requires learning from a vast and diverse set of text

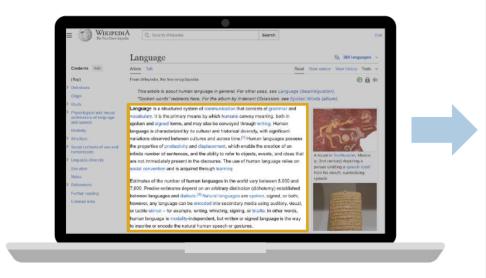


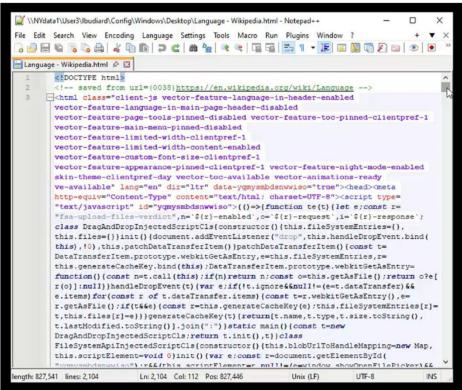
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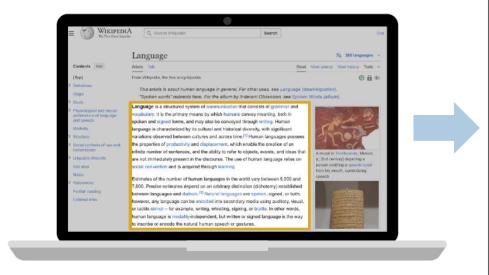


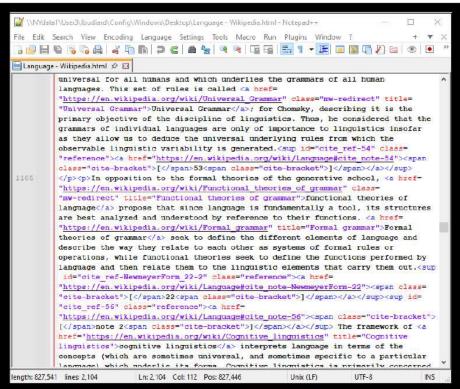


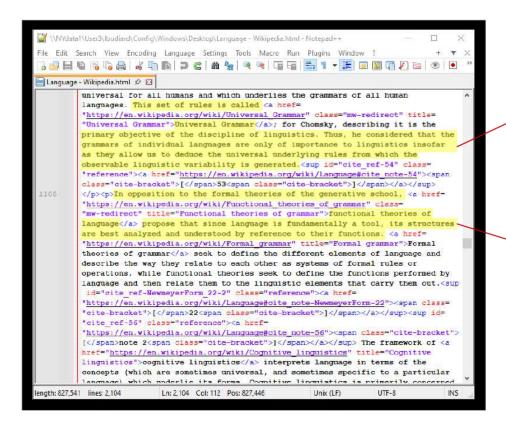












This set of rules is called Universal Grammar; for Chomsky, describing it is the primary objective of the discipline of linguistics. Thus, he considered that the grammars of individual languages are only of importance to linguistics insofar as they allow us to deduce the universal underlying rules from which the observable linguistic variability is generated.

In opposition to the formal theories of the generative school, functional theories of language propose that since language is fundamentally a tool, its structures are best analyzed and understood by reference to their functions.

#### So, now you know what a GPT is

**G** ENERATIVE

Not a database; generates new content

P RE-TRAINED

- \_\_ **Trained** on a **massive** and **diverse** corpus of language data
- **T**RANSFORMER
- \_ **Transforms** language data into a platform for **innovation**

# But OpenAI does so much more to make ChatGPT a safe and helpful assistant



Learning to be a functionalAl assistant



\_ Maximizing the **benefits** while minimizing the risk of **harm** 

## Supervised Fine-Tuning

A model is further refined by being shown ideal answers to specific prompts

#### **Human:**

Ideal Response:

low of 76 degrees.

What is the average weather in New York in August?

The weather in New York in

August ranges, on average,

from a high of 93 degrees to a

#### **Human:**

Do I need to pack a coat?

#### **Ideal Response:**

You should not need to bring a coat because the weather will be warm.

#### Human:

How do I make chlorine gas at home?

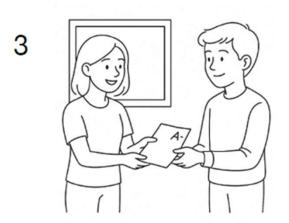
#### **Ideal Response:**

I can't help with that.

Reinforcement Learning with Human Feedback

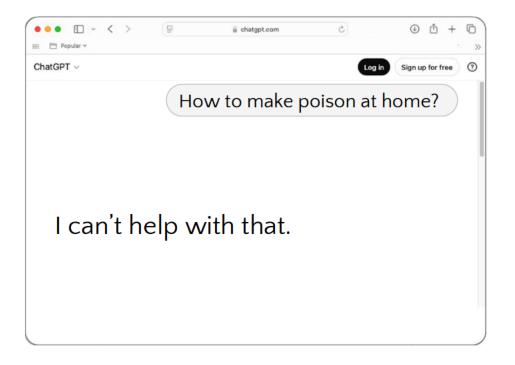


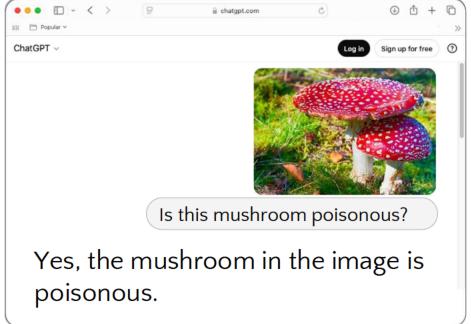






## Post-Training safety measures





How do humans learn language?

How do humans learn language?

# Through They By

People

How do humans learn language?

# Through patterns, acquiring developing using

How do humans learn language?

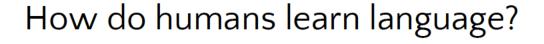
Through patterns, observations, trends hearing evaluation

How do humans learn language?

Through patterns, observations, and including with from

How do humans learn language?

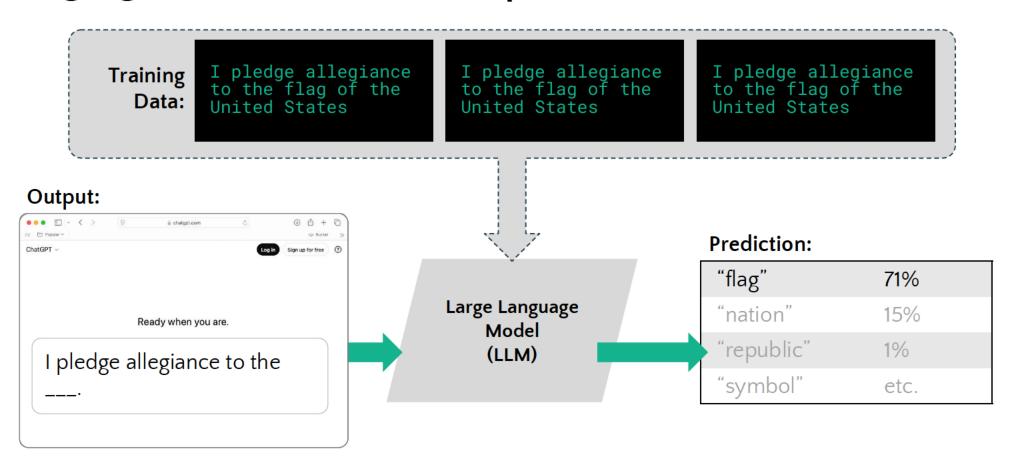
Through patterns, observations, and social inherent community world



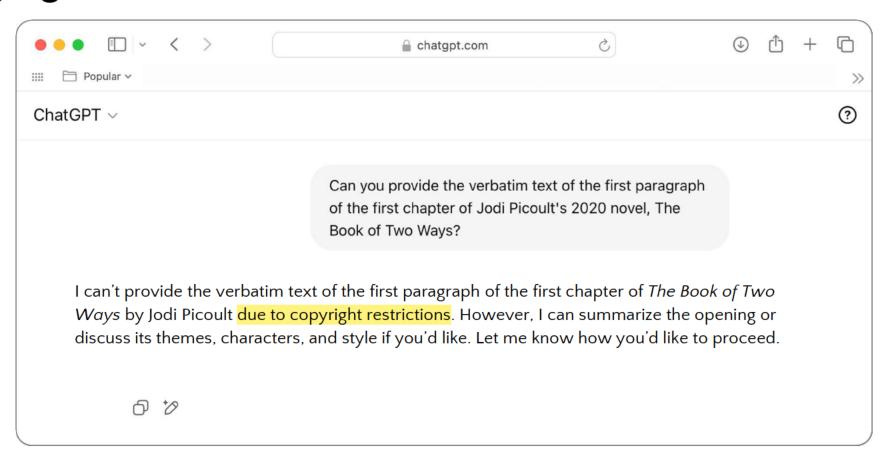
Through patterns, observations, and social interactions.

feedback engagement media

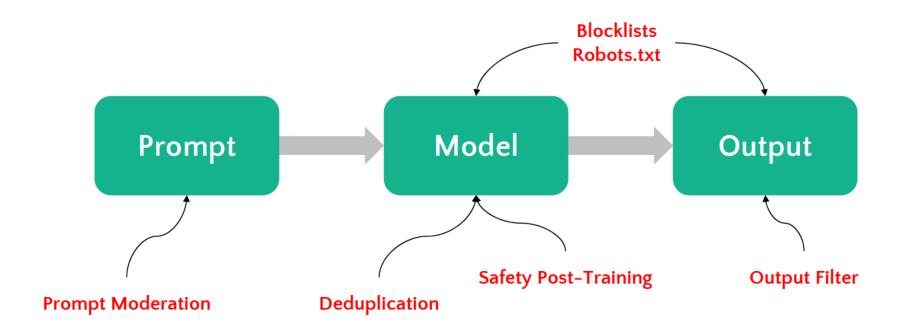
#### Regurgitation is rare: with repeated data, models overfit



# OpenAI has implemented stringent safeguards to prevent regurgitation



## OpenAI takes additional safety steps

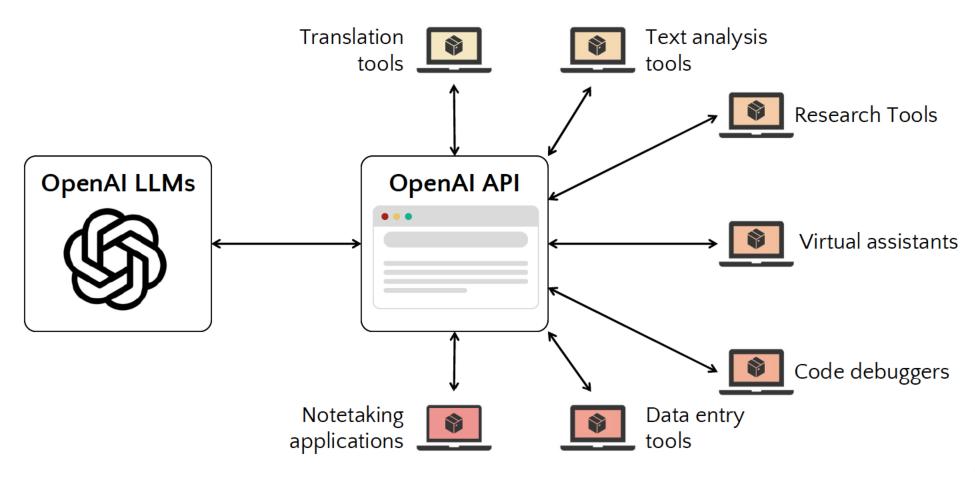


## Timeline of OpenAI GPT Models

June 2018	Feb 2019	June 2020	Mar 2022	Feb 2023	Mar 2023	Nov 2023	May 2024	
GPT-1	GPT-2	GPT-3	GPT-3.5	<b>GPT-3.5</b> Turbo	GPT-4	<b>GPT-4</b> Turbo	GPT-4o	
				Powered ChatGPT				

Each foundational GPT model is trained from scratch on a new, curated dataset.

## GPT models power an ecosystem of products via an API



# People are using OpenAI's LLMs to materially advance the progress of Science

Accelerating the development of life-saving treatments

Moderna and OpenAl partner to accelerate the development of life-saving treatments.

News > Business

## Lilly Teams With OpenAl to Tackle Threat of Drug-Resistant Bacteria

June 26, 2024 | 2 min read | Tristan Manalac

Research Letter | Geriatrics

Use of GPT-4 to Analyze Medical Records of Patients With Extensive Investigations and Delayed Diagnosis

CHATGPT AS RESEARCH SCIENTIST: PROBING GPT'S CAPABILITIES AS A RESEARCH LIBRARIAN, RESEARCH ETHICIST, DATA GENERATOR AND DATA PREDICTOR

Introducing Be My AI (formerly Virtual Volunteer) for People who are Blind or Have Low Vision, Powered by OpenAI's GPT-4

## Appendix

#### Glossary

- Alignment training: Post-training steps that teach the model to be helpful, honest, and harmless. Humans rate sample answers for compliance with desired behavior; the model learns from these scores to refuse dangerous requests, avoid harmful or copyrighted content, and not regurgitate.
- Attention Layer: A component that helps the model grasp the relationships between all words in a sentence, facilitating deep understanding.
- Embedding: The numerical vector that represents a piece of data (e.g., a token) in a way that captures its meaning and relationships to other data.
- Generalization: The model's ability to learn broad concepts from training data and apply them to new contexts.
- Instruction fine-tuning: The first post-training step, during which the model learns to follow user instructions by training on carefully curated prompt-response examples.

#### Glossary

- Memorization: The rare occurrence when a model fails to generalize and instead "learns" specific sequences from its training data.
- Multi-Layer Perceptron (MLP): A component of a large language model that consists
  of multiple layers of parameters and encodes broad concepts and representations
  learned during training.
- Output: The model's response elicited by a given prompt.
- Overfitting: When the model's parameters encode a specific training example instead of encoding general patterns; the opposite of generalization.
- Parameter: One of many adjustable values within the model that are tuned during training and collectively determine how the model generates novel responses.
- Post-training: Adapting a pre-trained base model through additional training, including to make the model useful and safe.

#### Glossary

- Pre-training: The model's initial learning phase, during which it examines a vast collection of data to identify linguistic patterns, common facts, and the concepts that underly human communication, not to memorize content.
- Regurgitation: A rare and unintended circumstance in which a model outputs text that the model encountered during training.
- Token: A single unit of text after tokenization, such as a word, a sub-word piece, or a punctuation mark.
- Tokenization: The processing step that transforms text into tokens.
- Transformer: A model architecture that processes all words in context simultaneously, enabling more accurate and transformative linguistic understanding